

An Introduction to Process Variability Simulation

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Abstract—The process variability covers typically the material, equipment, method, and man, the so-called 4 Ms'. A simple process of cutting a 10-meter rod into two equal halves is illustrated where all the 4Ms were treated. The simulation of variability for the 4 Ms' were discussed. The theoretical perspective has caveat on matter of mathematical expressions that are realistic. The algorithm used was Monte Carlo Simulation based on random number generation that fit Gaussian Curve. The python programming language was used with symbolic math library, the sympy. Although the industry is used to 99.99%, for the sake of the academy around 50%.

Index Terms—variability, Normal Distribution, Gaussian, Man, Machine, Method, Material, 4Ms.

1 INTRODUCTION

The Monte Carlo Simulation requires the generation of random variable of which statistical distribution follows a certain distribution. In this case, the normal or Gaussian distribution is used based on the assertion that a stable process must obey central tendency theorem.

The Gaussian distribution function [4] is stated as follows.

$$f(x) = \frac{\sqrt{2}e^{-\frac{(\mu-x)^2}{2\sigma^2}}}{2\sqrt{\pi}\sigma}$$

where :

μ : Real number or a list representing the mean or the mean vector

σ : Real number or a positive definite square matrix,
 $\sigma^2 > 0$ the variance

Returns: A Random Symbol.

The sympy function

$$f_1 = \text{NormalDistribution}(\mu, \sigma)$$

declares f_1 as the normal distribution function. The random variable of z of it is expressed as follows.

$$f_1(z) = \frac{\sqrt{2}e^{-\frac{(\mu-z)^2}{2\sigma^2}}}{2\sqrt{\pi}\sigma} \quad (3)$$

Integrating (3), we have the following.

$$f_2(z) = \int_{z_i}^{z_f} f_1(z) dz = \frac{\text{erf}\left(\frac{\sqrt{2}(-\mu+z_f)}{2\sigma}\right)}{2} - \frac{\text{erf}\left(\frac{\sqrt{2}(-\mu+z_i)}{2\sigma}\right)}{2} \quad (4)$$

Let $\mu = 0$ and $\sigma = 1$ then f_2 from $z_i = -\infty$ to $z_f = \infty$ we have normal probability distribution as follows.

$$f_2(z) = \int_{-\infty}^{\infty} f_1(z) dz = 1 \quad (5)$$

The $f_3(z)$ from $z_i = -1$ to $z_f = 1$ implies lower limit at $-\sigma$ and upper limit σ .

$$\int_{-1}^1 f_1(z) dz = \text{erf}\left(\frac{\sqrt{2}}{2}\right) = 0.682689492137086 \quad (6)$$

At $\pm 3\sigma$,

$$\int_{-3}^3 f_1(z) dz = \text{erf}\left(\frac{3\sqrt{2}}{2}\right) = 0.99730020393674 \quad (7)$$

Finally at $\pm 6\sigma$,

$$\int_{-6}^6 f_1(z) dz = \text{erf}\left(3\sqrt{2}\right) = 0.999999998026825 \quad (8)$$

The defect per million DPM [1] of (6), (7), and (8) are given as follows.

$$DPM = 317310.507862914 \quad (9)$$

$$DPM = 2699.79606326021 \quad (10)$$

$$DPM = 0.00197317528982666 \quad (11)$$

2 REVIEW OF RELATED LITERATURE

The context of this tutorial on process variability is the statistical process control (SPC) [2]. It is asserted that the stable process follows the central limit theorem [5]. Hence, by making the upper control limit and lower control limit of approaches a distant of 6σ from the mean μ , the process is made extremely stable and predictable. The discussion would focus on the relationship of variables in processes to account for relative impact of assignable causes [3].

3 SIMULATION

Given a case where a 10m rod is to be cut into two halves where the statistics of random variables are those of machine, man, material and method, it is desired to find out how these affect the yield of the process. The objective of the analysis is to improve the accuracy that is the distance of average from the desired mean and the precision that is standard deviation or variance.

3.1 Material

A set of values of z can be generated randomly by the `sympy` function $Normal(\mu, \sigma)$. Given $\mu = 10$ and $\sigma = 1$ then for 10000 pieces we have

$$\begin{aligned} \text{Rods} &= \text{random.normal}(\text{nrodmu}, \text{nrodsigma}, \text{count}) \\ &= [8.866765, 8.808836 \dots 8.053214]. \end{aligned} \quad (12)$$

The histogram of (15) is shown in Figure 1.

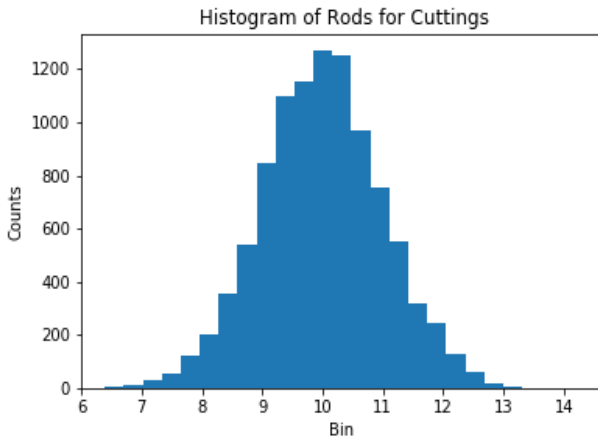


Fig. 1. 1. Histogram Rods Variation

Question:

Given the specification of ± 1 , how many are good and what is the yield?

Answer number of good rods = 6954, yield = 69.54%. Note that everytime the program is run these values changes. The solution was an algorithm to check dimension of each

rod and count those that are within tolerance. Solving it using (4) gives a unique answer unlike the results from the algorithm above. Why?

3.2 Equipment

Now consider that these rods are to be cut into halves. The cutter has $\mu = 0.5$ and $\sigma = 0.1$ then for 6954 pieces we have

$$f_1 = \text{NormalDistribution}(0.5, 0.1) \quad (13)$$

$$f_1(z) = \frac{5.0\sqrt{2}e^{-50.0(z-0.5)^2}}{\sqrt{\pi}} \quad (14)$$

$$f_2(z) = \int_{z_i}^{z_f} f_1(z) dz \quad (15)$$

$$\begin{aligned} \text{Cut}_{event} &= \text{random.normal}(0.5, 0.1, 6954) = \\ &= [0.5104791588224948, 0.5032845297161098, \dots \\ &0.4233040541115318] \end{aligned} \quad (16)$$

The histogram of (19) is shown in Figure 2.

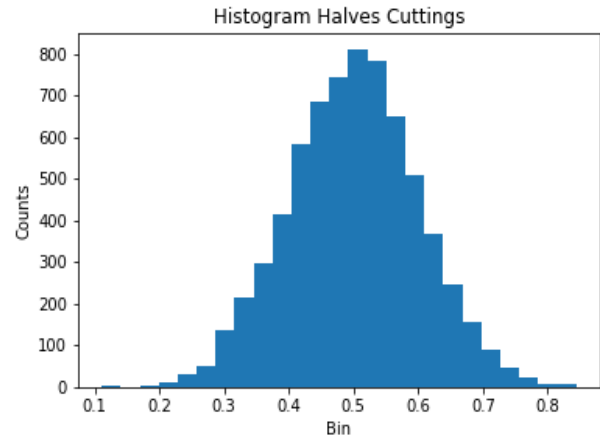


Fig. 2. Histogram of Cutter Variation

3.3 Method

There two methods to cut the rod. One method is the cutter is place at a distance of half the rod size exactly 5. Then the rod is referenced at the point of origin and cut is made. The other method is to place the rod at its center for cutter engagement. The formulas for the two methods are different.

Question: Determine the formula for each and justify it.

Answers:

1. The cutter with variation, i.e., 0.5 ± 0.1 is multiplied by the perfect dimension of rod that is 10. Why perfect 10? Then the answer is subtracted from a rod with variation. The two answers are the cut products.

2. The rod with variation, i.e., 10 ± 1 is multiplied by the cutter with variation, i.e., 0.5 ± 0.1 . The answer is subtracted from the rod with variation. The two cuts are products. The histograms of Method 1 and 2 are shown in Figure 3.

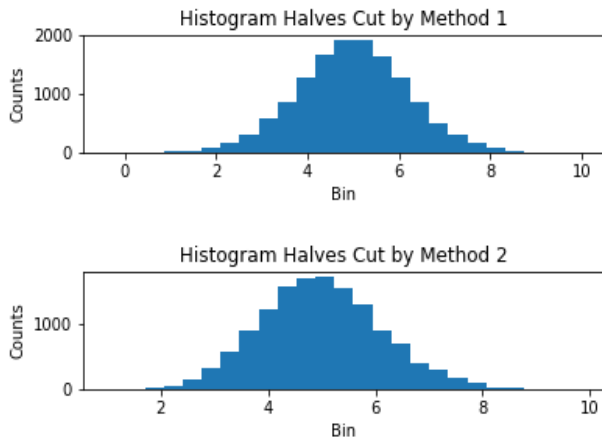


Fig. 3. Histogram of Cut Rods

Question:

Assuming that the cut specification is $5 \pm .5$, determine the number of good cut and the yield for each method

Answer:

1. The number of good cut for method 1 is 4557 while for method 2 4780 respectively.
2. The yield for method 1 is 32.7653% while for method 2 is 34.3687% respectively. Why is it method 2 has higher yield?

3.4 Operator

Assume an operator makes the loading with error of $\mu = 0.4$ and $\sigma = 0.1$. The histograms with operator handling for Method 1 and 2 are shown in Figure 4.

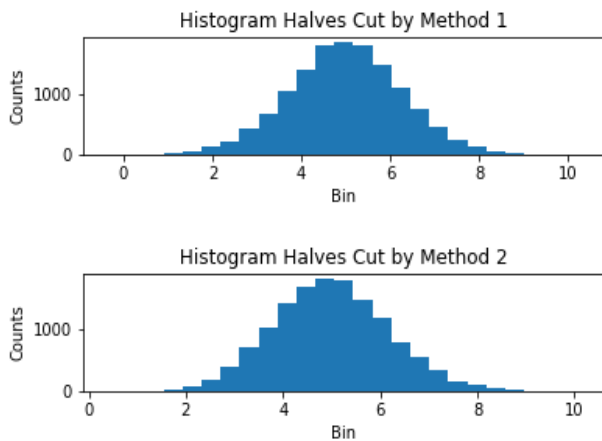


Fig. 4. Histogram of Cut Rods with Operator's Handling

The yield with operator's handling errors for method 1 is 31% while for method 2 is 32% respectively.

4 DISCUSSION

4.1 Tolerance Variation

The yield numbers were exaggerated for theoretical discussion. The four variables are material, equipment

or machine, method, and operator or man. Hence, such sources of variations are known as 4M.

In real world, the overall yield is the only one observed and reported. It is problematic to account the yield loss for each source. However, the so-called fool proofing refers to elimination of operator's mishandling. Typically the solution is jig and fixture. In this case, the positional fixture with alignment jig could eliminate the operator's error.

The variation in machine could increase overtime. Hence, regular calibration is necessary. To improve the variability, the continuous improvement program is the typical solution.

The material variation can be mitigated at a cost by simple requirement for tighter tolerances. However, the most difficult variation to estimate is the method. The difficulty was choosing the appropriate equation to define the variability in a method. For instance, either of the two methods has rejects of oversize. It is possible to introduce rework or reprocess to cut the oversize for yield recovery purposes. What are the appropriate equations and the corresponding algorithm that can be concocted for each method? If the target reference remains the mean, there is a chance that the outcome is undersize. Should the reference be at maximum tolerance? In other words, the method for rework may not be the same as the standard process.

4.2 Other Type of Variations

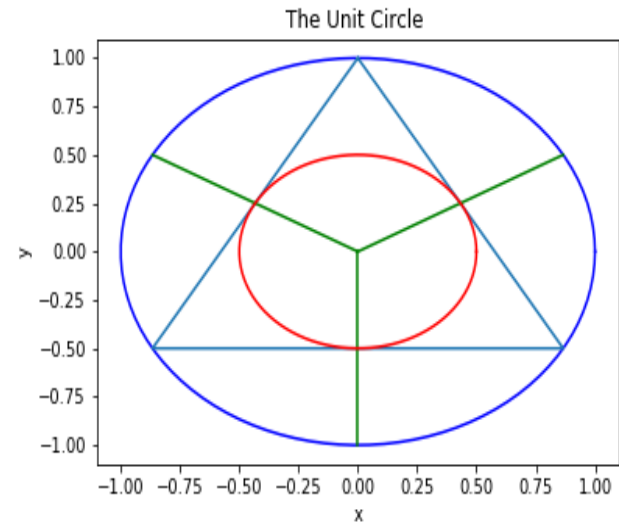


Fig. 5. Equilateral Triangle Inscribed in Unit Circle

Consider a unit circle where an equilateral triangle is inscribed as shown in Figure 5. Suppose a random line is to be drawn across it, determine the probability that the cord is less than the length of the side of the equilateral triangle. There are a number of answers depending on how the random lines are drawn.

1. The line drawn from any two points at the circumference,

gives the probability of $\frac{1}{3}$.

2. The line drawn perpendicular to radius from a any point along radius gives the probability of $\frac{1}{2}$
3. The line drawn tangent to any concentric circle gives the probability $\frac{1}{4}$.
4. From the length of the cord i.e. 0 to 2, the probability is $\frac{\sqrt{3}}{2}$.

The question is that which of the four probability numbers make sense?

5 CONCLUSION

The variation of process variables can simulated by pseudo random generator. The formula or equation associated with the algorithm must be ensured realistically. From theoretical perspective the variance of every variable can be determined but in real world operation, the net variance is the only one that can be measured. Hence, the defect modes may be difficult to itemize.

As discussed, the application of random variable depends on certain perspective of variation. For instance, a short cord drawn near the center of the circuit is not a line crossing the circle although it is random. It hsd to be moved from the center to have its end points touch the circumference.

Finally, the accuracy must be the difference between the average and the specified mean must be approaching zero. The precision or repeatability must be such that the tolerance is equal to a number of sixmas.

6 RECOMMENDATION

The Monte Carlo Simulation is useful for prediction of process yield. Due care should be excercised in formulation of equation and algorithm. The mathematics must be articulate the real world. There exists ample room for creativity in defect modes itemization. The design of experiments were used to optimize parameters. The Monte Carlo Simulation may help to reduce number of experiments.

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